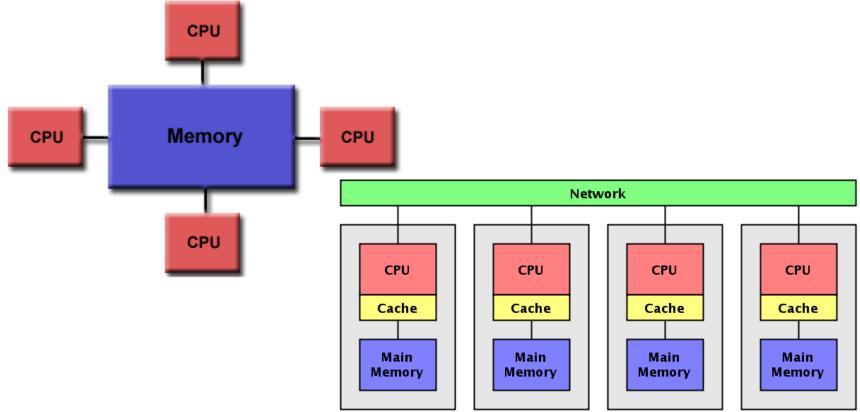
Introduction to Parallel Computing

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What is Parallel Computing?

- Wikipedia says: "Parallel computing is a form of computation in which many calculations are carried out simultaneously"
- Speed measured in FLOPS

What is Parallel Computing? (cont.)



How can this be useful?

- Faster Calculations
- Make use of less powerful hardware?
- Many computers, or "nodes" can be combined into a cluster
- But, it's a lot more complex to implement

Definitions

- Core: A processor that carries out instructions sequentially. Higher frequency means faster calculations
- Process/Task: A chain of instructions; a program

Main Granularity Paradigms

Granularity: the ratio of computation to communication

 3 approaches to parallelism, depending on what kind of problem you need to solve

Embarrassingly Parallel

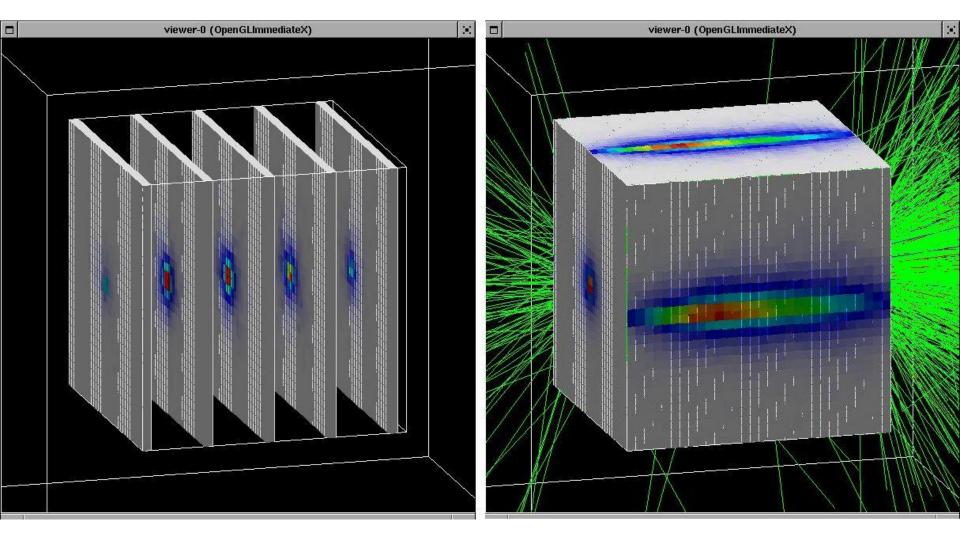
- No effort required to separate tasks
- Tasks do not depend on, or communicate with, each other.
- Examples: Mandlebrot, Folding@home,
 Password brute-forcing, Bitcoin Mining!

Example: Particle Physics

Independent tracked particles

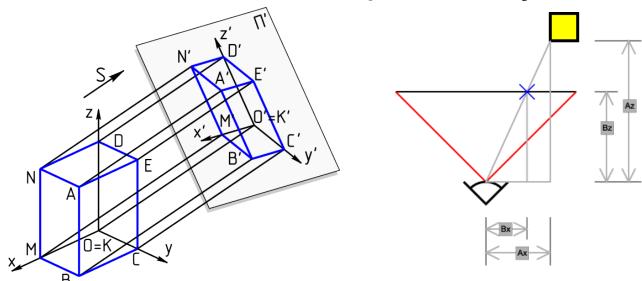
They interact with the world, not each other

Can be submitted as a batch of jobs



Example: 3D Projection

 Each pixel on the screen, or each block of pixels, is rendered independently



Coarse-grained Parallelism

- Tasks communicate with each other, but not more that once a second
- Examples: Stuff that involves synchronization

Fine-grained Parallelism

AKA Multithreading

Subtasks must constantly communicate with each other

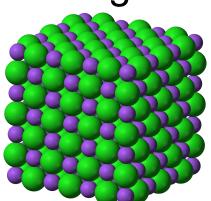
Must use something like MPI

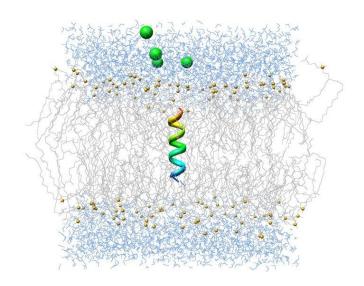
Example: Molecular Dynamics

Relaxation of a protein in water

Movement of atoms depends on that of

surrounding atoms





Job Scheduling

- Integral to parallel computing; assigns tasks to cores
- Batch jobs, Multiple users, Resource sharing, System monitoring

Livelock/Deadlock/Race Conditions

- Things that could go wrong when you are performing a fine or coarse-grained computation:
- Livelock
- Deadlock
- Race Conditions

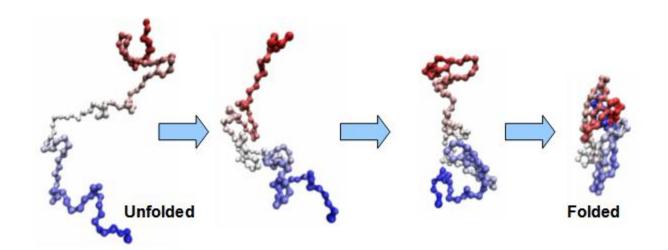
Not Everything Benefits

 Many problems must be solved sequentially (e.g. Long division, protein folding)

Interactive stuff

Example: Protein Folding

Each atom depends on the one before it



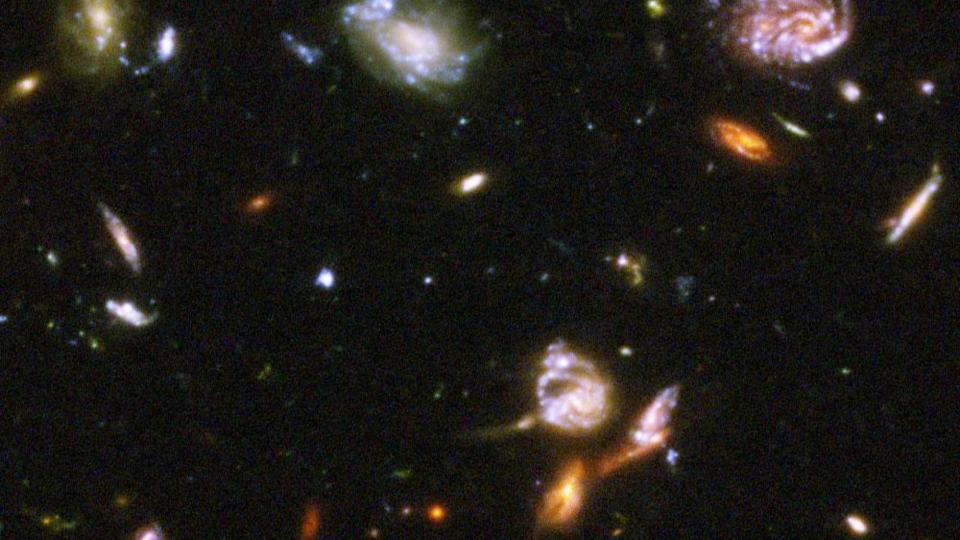
Slowdown, Bottlenecks

- How fast can cores communicate with each other?
- How fast can nodes communicate with each other?
- Infinitiband

Massively parallel computing

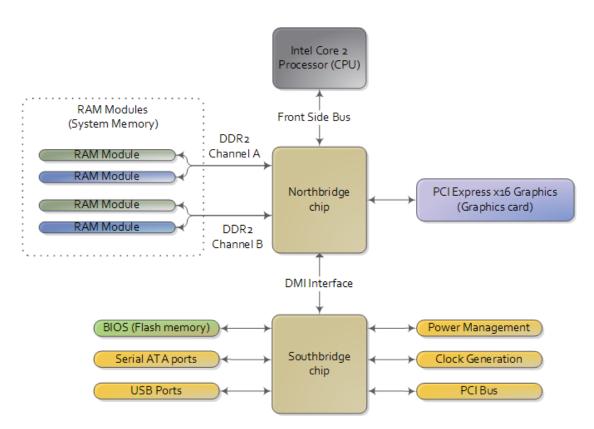
Clusters, Supercomputers

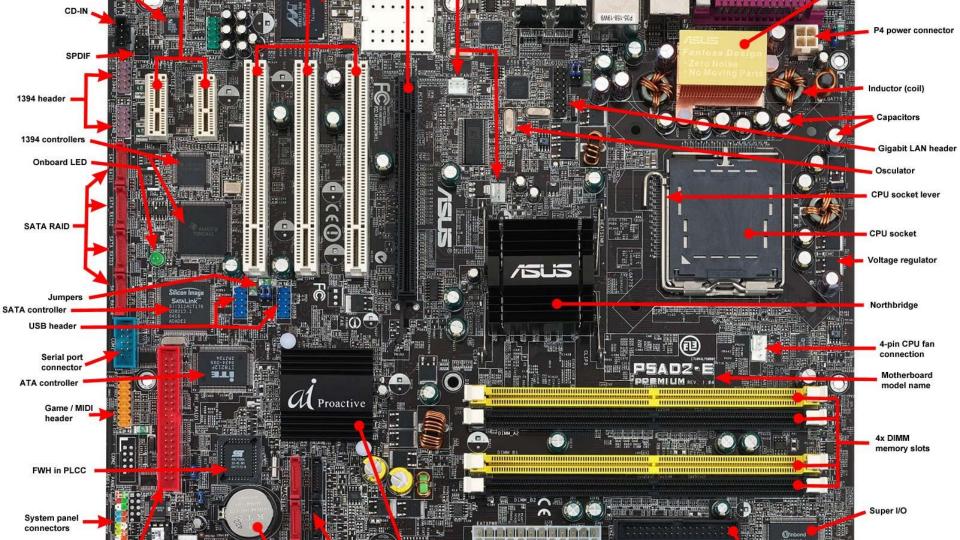
 Grids (Folding@home, SETI@home, PlanetQuest)



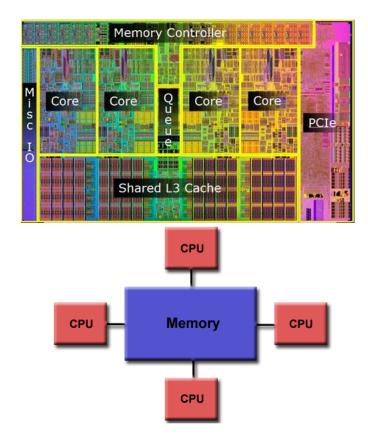
CPU Methods

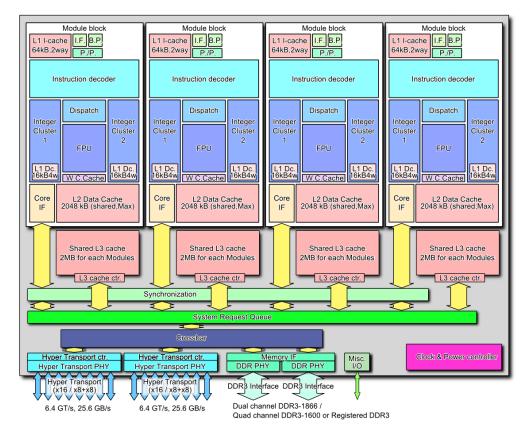
Computer Overview





CPU Architecture





CPUs can't get much faster!

- Temperature
- Lithography limitations
- Quantum tunneling
- Electricity travel speed
- We can add more cores though

Message Passing Interface

 Allows individual processes to talk to processes on different cores

MPI Implementations

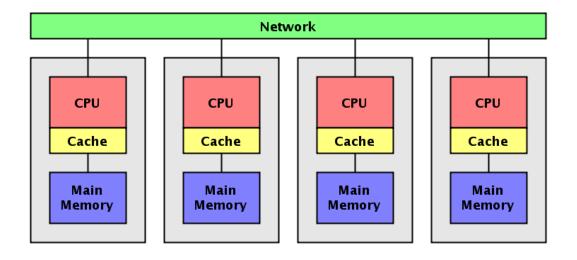
- OpenMPI, MPICH
- Wrappers for various languages
- mpirun -np process-count program-name

Example Code

```
#define NPTS 1000000000
#define NLOOP 10
void main(int argc, char** argv) {
    int rank, nproc, i, istart, iend, loop, N;
   unsigned long int start time, end time;
   struct timeval start, end;
    double sum, pi, mflops;
    MPI Init (&argc, &argv);
    MPI Comm rank (MPI COMM WORLD, &rank);
    MPI Comm size (MPI COMM WORLD, &nproc);
    if (rank == 0) {
        gettimeofday(&start, NULL);
        istart = 1 + NPTS*((rank+0.0)/nproc);
        iend = NPTS*((rank+1.0)/nproc);
        sum = 0.0;
        for (loop = 0; loop < NLOOP; ++loop)
         for (i = istart; i <= iend; ++i)
                 sum += 0.5/((i-0.75)*(i-0.25));
        MPI Reduce (&sum, &pi, 1, MPI DOUBLE, MPI SUM, 0, MPI COMM WORLD);
        gettimeofday(&end, NULL);
         start time = start.tv sec * 1e6 + start.tv usec;
        end time = end.tv sec * 1e6 + end.tv usec;
        mflops = NLOOP*(NPTS*(5.0/(end time-start time)));
        printf("processes = %d, NPTS = %d, NLOOP = %d, pi = %f\n", nproc, NPTS, NLOOP,
pi/NLOOP);
        printf("time = %f, estimated MFlops = %f\n", (end time-start time) /1.0e6, mflops);
   else
        istart = 1 + NPTS*((rank+0.0)/nproc);
        iend = NPTS*((rank+1.0)/nproc);
        sum = 0.0;
        for (loop = 0; loop < NLOOP; ++loop)
         for (i = istart; i <= iend; ++i)
                 sum += 0.5/((i-0.75)*(i-0.25));
        MPI Reduce (&sum, &pi, 1, MPI DOUBLE, MPI SUM, 0, MPI COMM WORLD);
    MPI Finalize();
```

Expanding Into a Cluster

 Multiple computers can be linked together over a network. MPI can use this



GPU Methods

Why were GPUs created?

- Simply, wanted to free up CPU
- GUIs required programmers to think in different ways
- In a GUI, everything behaves independently

GPU Architecture

- Like a multi-core CPU, but with thousands of cores
- Has its own memory to calculate with



GPU Advantages

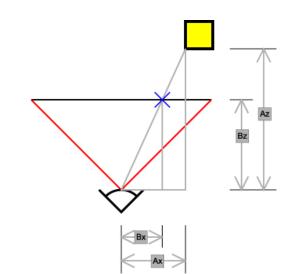
- Ridiculously higher net computation power than CPUs
- Can be thousands of simultaneous calculations
- Pretty cheap

Historic GPU Programming

- First developed to copy bitmaps around
- OpenGL, DirectX
- These APIs simplified making 3D games/visualizations

Pipeline for rendering 3D

- Vertex data sent in by graphics API (from CPU code via OpenGL or DirectX, for example)
- Processed by vertex program (shader)
- Rasterized into pixels
- Processed by fragment shader

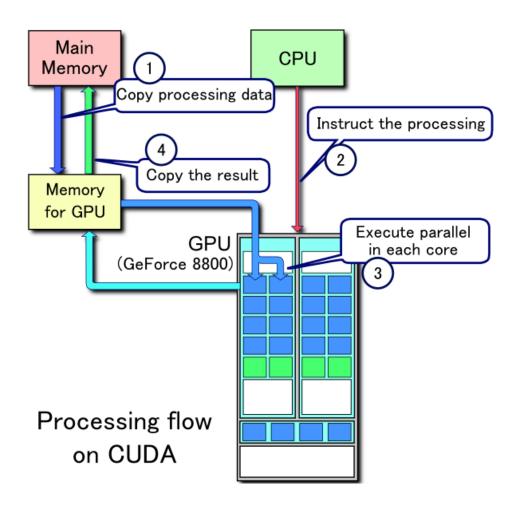


Modern GPU Frameworks

 CUDA: Proprietary, Easy to user, Sponsored by NVIDIA and only runs on their cards

 OpenCL: Open, a bit harder to use, runs on both ATI and NVIDIA arch

CUDA Methods



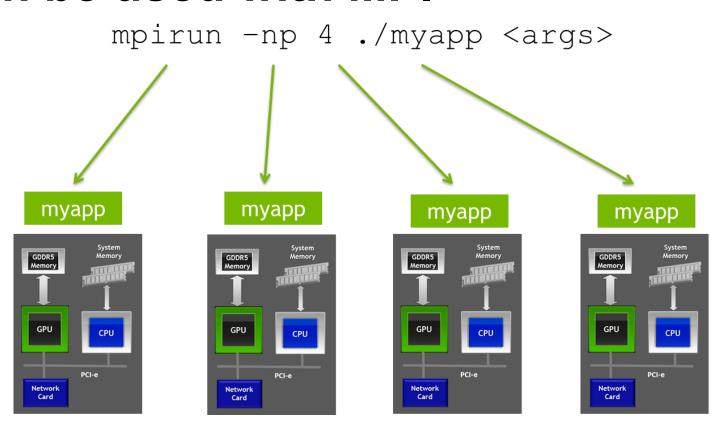
CUDA Example Code

Standard C Code

C with CUDA extensions

```
//Vector size in elements
const int N = 1048576:
//Vector size in bytes
const int dataSize = N * sizeof(float);
//CPU memory allocation
float *h A = (float *)malloc(dataSize);
float *h B = (float *)malloc(dataSize);
float *h C = (float *)malloc(dataSize);
//GPU memory allocation
float *d A, *d B, *d C;
cudaMalloc((void **)&d A, dataSize));
cudaMalloc((void **)&d B, dataSize));
cudaMalloc((void **)&d C, dataSize));
//Initialize h A[], h B[]...
//Copy input data to GPU for processing
cudaMemcpy(d A, h A, dataSize, cudaMemcpyHostToDevice) );
cudaMemcpy(d B, h B, dataSize, cudaMemcpyHostToDevice) );
//Run the core of N / 256 units, 256 streams each
//Assuming that N is multiple of 256
vectorAdd <<<N / 256, 256>>>(d C, d A, d B);
//Read GPU results
cudaMemcpy(h C, d C, dataSize, cudaMemcpyDeviceToHost) );
```

Can be used with MPI



How to Setup Your Own Parallel Computer

CPU or GPU Based?

 CPU: Easier to program for, has much more powerful individual cores

 GPU: Trickier to program for, thousands of really weak cores

Cluster or Multicore?

- Multicore: All cores on in a single computer, usually shared memory.
- Cluster: Many computers linked together, each with individual memory

OS/Distribution?

Updated or Stable?

Linux or something crazy?

Custom cluster distro?

Job Scheduler?

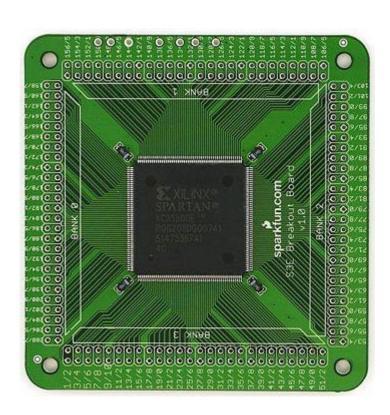
Submit stuff in batches or MPI?

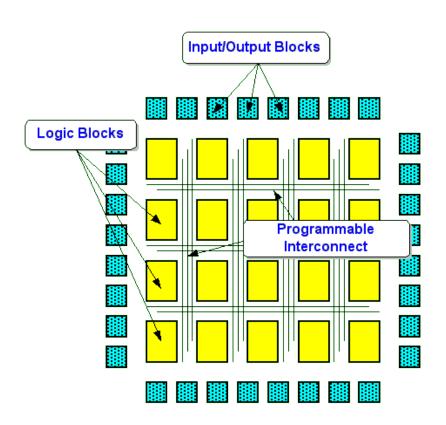
Grid Engine, Condor, SLURM

Programming It

Other Methods and their Applications

FPGAs





Cryptography

- numbers
- brute-forcing

Bitcoin

Proof-Of Work